

Mathematical modeling

Model

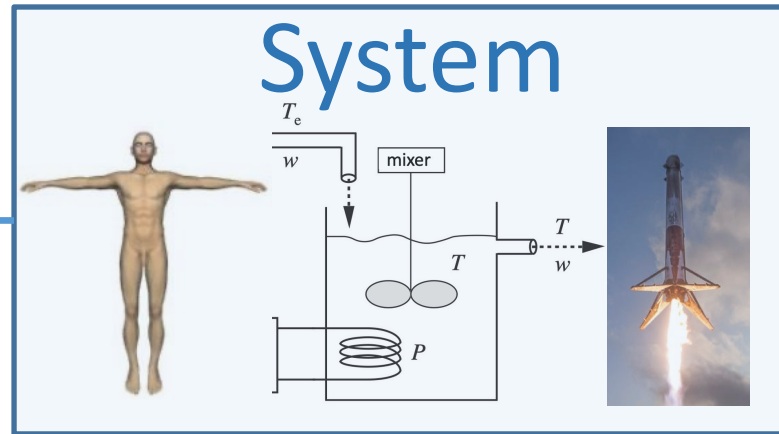
A less complex representation of the system to be studied

Why making a model?

- **Simulation and analysis**
 - gain an understanding of the behavior of the system
 - estimate a quantity or measure an objective
 - **Control**
 - **Optimization**
- } Evaluate alternative strategies

There is a trade-off between **model accuracy** and **complexity** and the **cost/effort** required to develop the model.

Classes of models



Known physical laws
Limited complexity

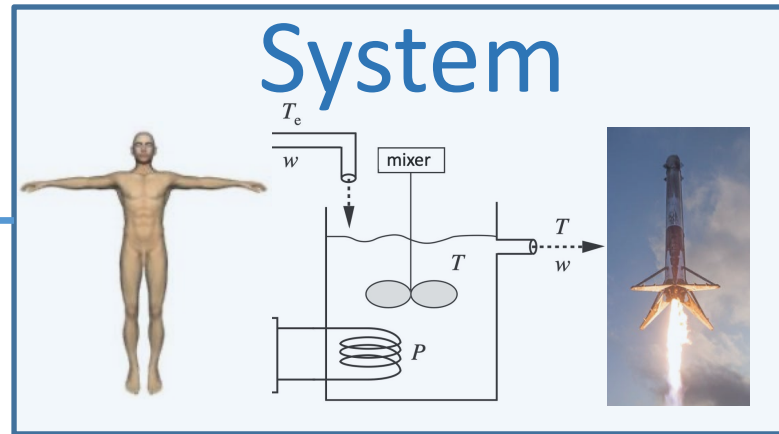
Analytically derived
KNOWLEDGE-BASED models

Molar balance

$$\frac{dC_A}{dt} = -kC_A$$

$$C_A(t_0) = C_{A0}$$

Classes of models



Known physical laws
Limited complexity

Observed behavior
(dominant or specific)

Analytically derived
KNOWLEDGE-BASED models

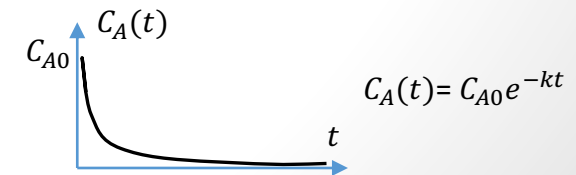
Experimentally derived
REPRESENTATION models

Molar balance

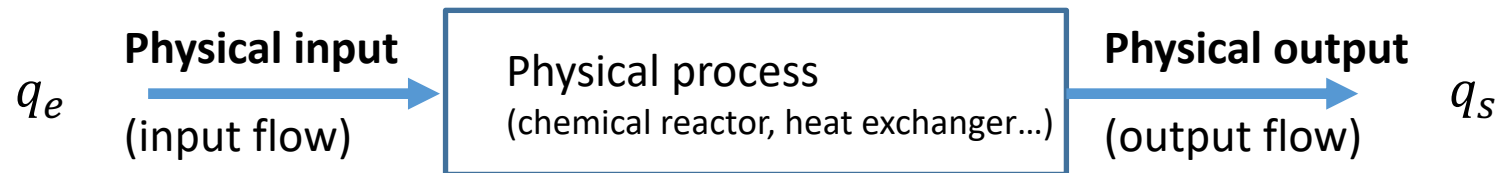
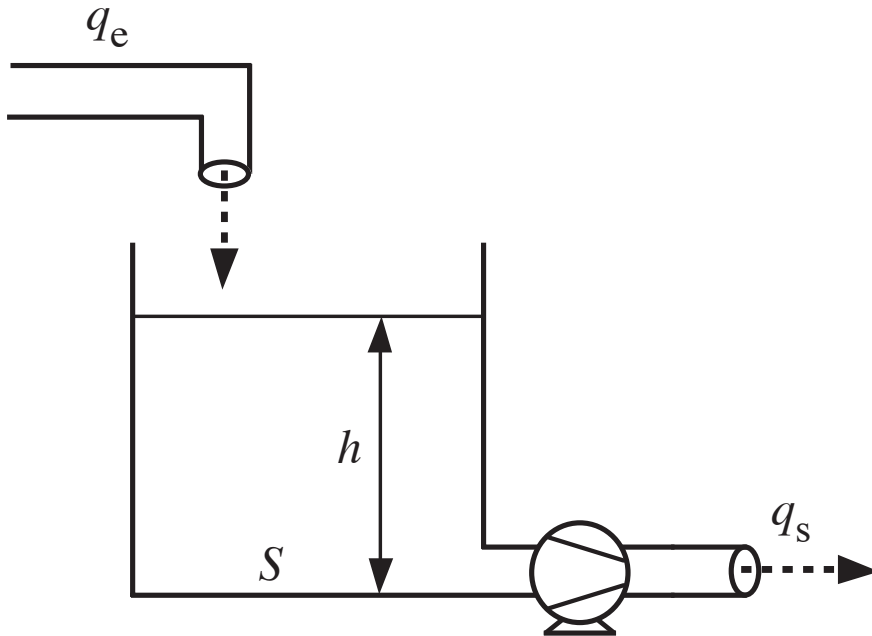
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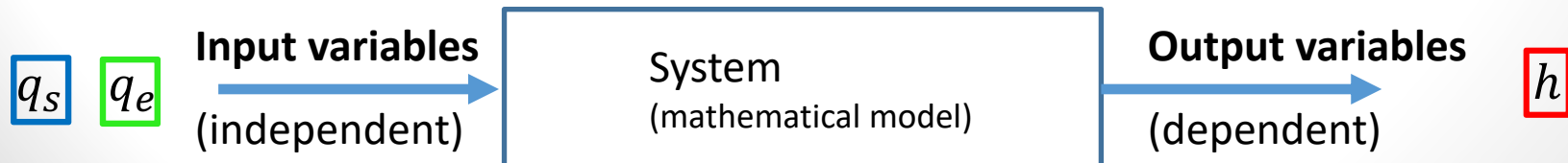
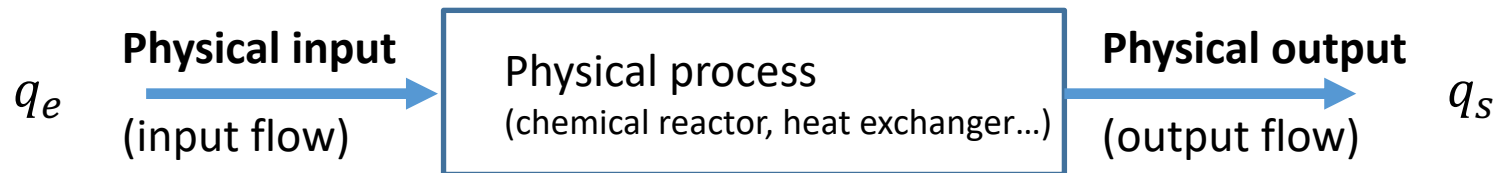
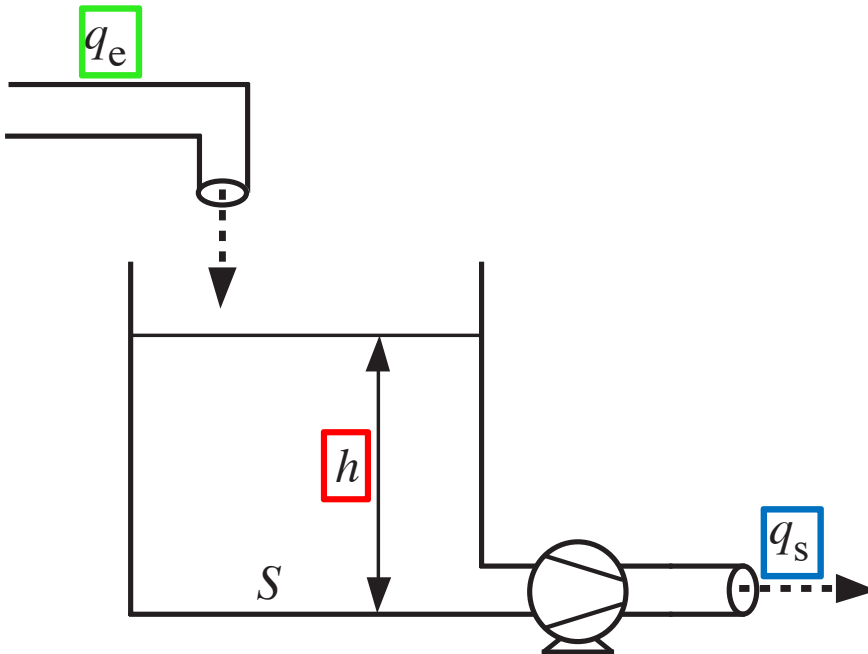
Observation



Input/output



Input/output



Stirred tank with constant volume

Assumptions

- V is constant
- No heat loss to the outside
- Well-mixed tank $T_s = T(t)$

Global mass balance

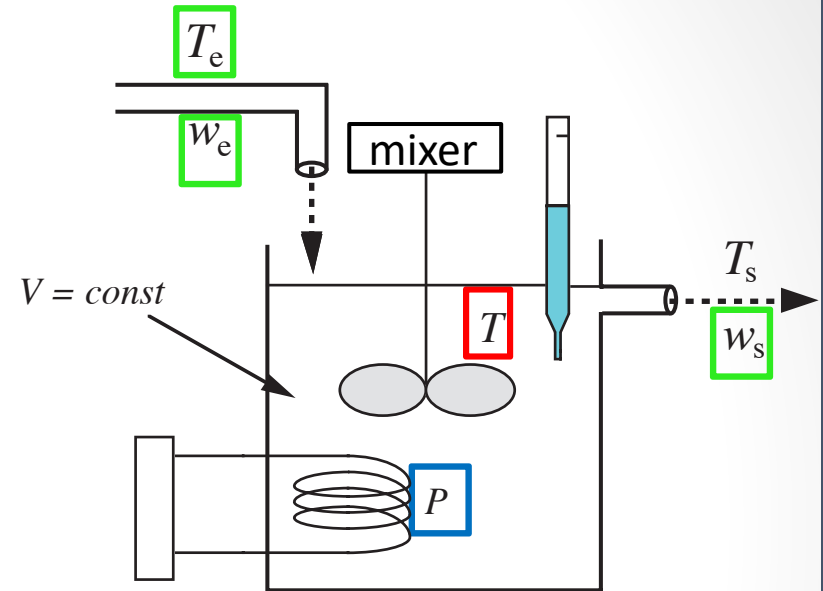
$$w(t) \stackrel{\text{def}}{=} w_s(t) = w_e(t)$$

Global energy balance

$$\left(\begin{array}{c} \text{rate of energy} \\ \text{accumulation} \end{array} \right) = \left(\begin{array}{c} \text{rate of} \\ \text{input energy} \end{array} \right) - \left(\begin{array}{c} \text{rate of} \\ \text{output energy} \end{array} \right)$$

$$\frac{d}{dt} [V \rho c_p (T(t) - T_{\text{ref}})] = [w(t) c_p (T_e(t) - T_{\text{ref}}) + P(t)] - w(t) c_p (T(t) - T_{\text{ref}})$$

$$V \rho c_p \frac{d}{dt} T(t) = w(t) c_p [T_e(t) - T(t)] + P(t)$$



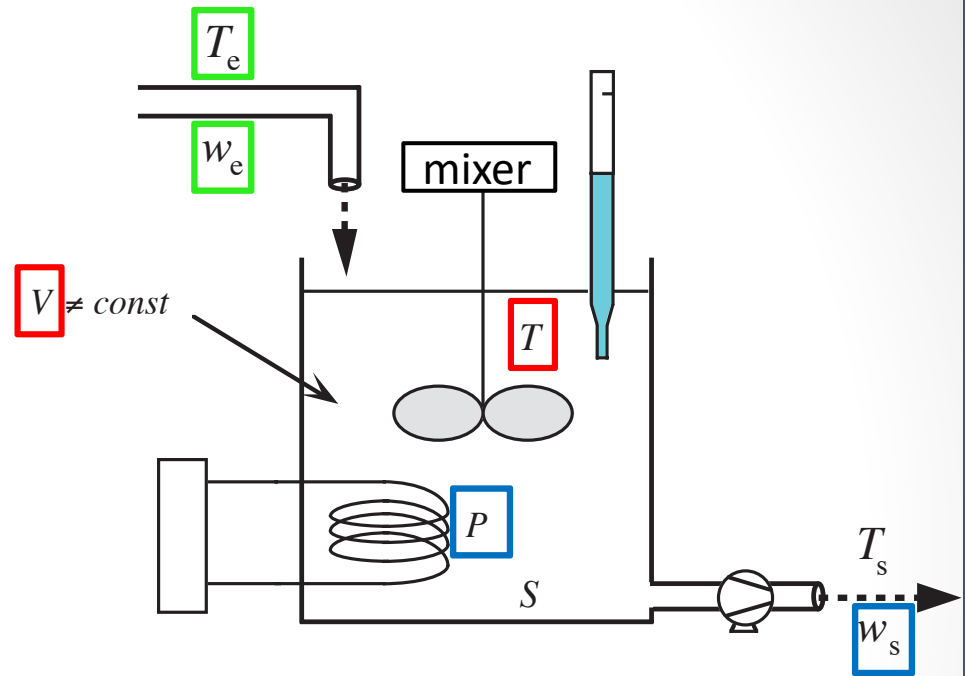
Stirred tank with variable volume

Assumptions

- V is **not** constant
- No heat loss to the outside
- Well-mixed tank $T_s = T(t)$

Global mass balance

$$w_s(t) \neq w_e(t)$$



$$\left(\begin{array}{c} \text{rate of mass} \\ \text{accumulation} \end{array} \right) = \left(\begin{array}{c} \text{input mass} \\ \text{flow rate} \end{array} \right) - \left(\begin{array}{c} \text{output mass} \\ \text{flow rate} \end{array} \right)$$

$$\frac{d}{dt} [\rho V(t)] = w_e(t) - w_s(t)$$

$$\rho \frac{d}{dt} V(t) = w_e(t) - w_s(t)$$

Stirred tank with variable volume (cont'd)

Global energy balance

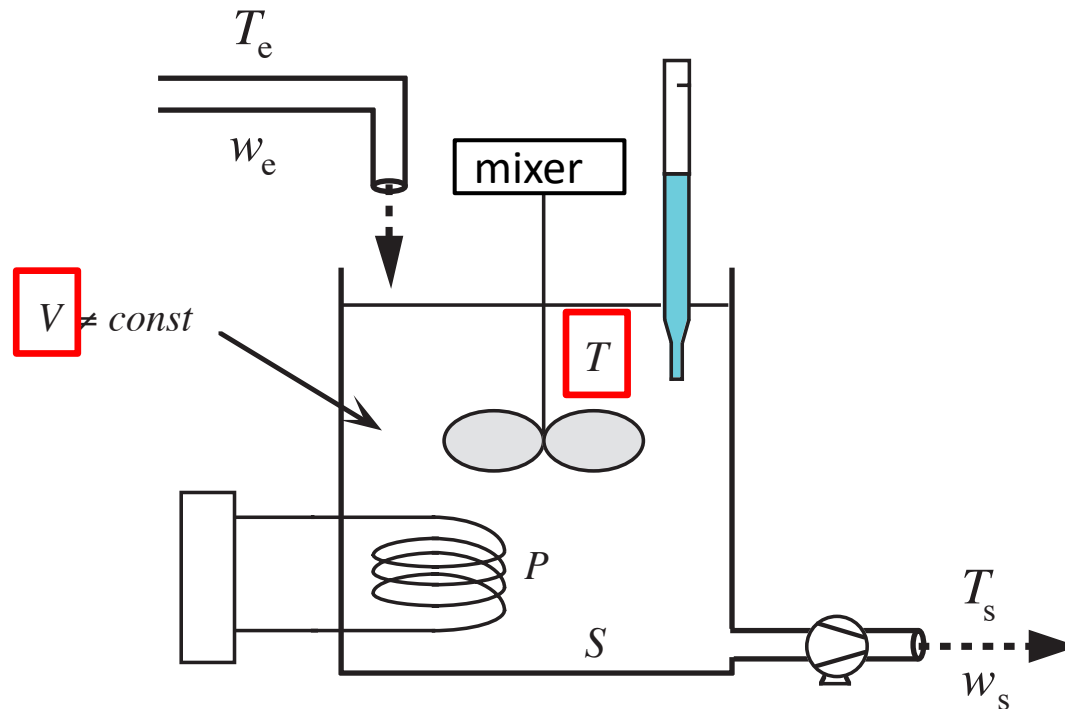
$$\left(\begin{array}{c} \text{rate of energy} \\ \text{accumulation} \end{array} \right) = \left(\begin{array}{c} \text{rate of} \\ \text{input energy} \end{array} \right) - \left(\begin{array}{c} \text{rate of} \\ \text{output energy} \end{array} \right)$$

$$\frac{d}{dt} [V(t) \rho c_p (T(t) - T_{\text{ref}})] = [w_e(t) c_p (T_e(t) - T_{\text{ref}}) + P(t)] - w_s(t) c_p (T(t) - T_{\text{ref}})$$

$$\rho \frac{d}{dt} V(t) = w_e(t) - w_s(t)$$

$$\frac{d}{dt} T(t) = \frac{w_e(t)}{V(t) \rho} [T_e(t) - T(t)] + \frac{P(t)}{V(t) \rho c_p}$$

Stirred tank with variable volume (cont'd)



$$\rho \frac{d}{dt} V(t) = w_e(t) - w_s(t)$$

$$\frac{d}{dt} T(t) = \frac{w_e(t)}{V(t)\rho} [T_e(t) - T(t)] + \frac{P(t)}{V(t)\rho c_p}$$

Stirred tank with variable volume (cont'd)

Specific cases

a) If V is constant:

$$\frac{d}{dt}V(t) = 0$$

$$w(t) := w_e(t) = w_s(t)$$

$$V\rho c_p \frac{d}{dt}T(t) = w(t)c_p [T_e(t) - T(t)] + P(t)$$

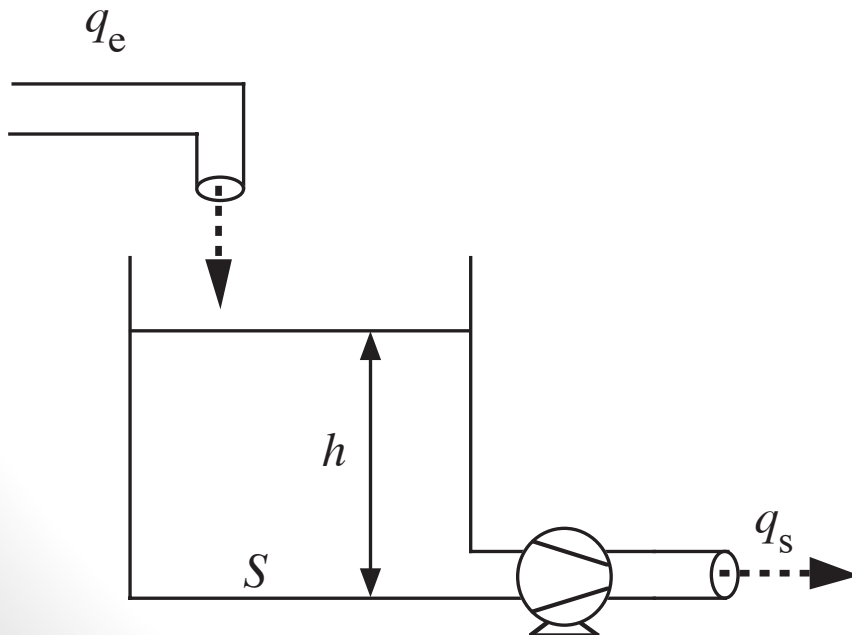
b) If T is constant:

Mass balance

$$\frac{d}{dt}[\rho V(t)] = \rho q_e(t) - \rho q_s(t)$$

ou

$$S \frac{d}{dt}h(t) = q_e(t) - q_s(t)$$



c) If V and T are constant (steady state):

$$\frac{d}{dt}V(t) = 0 \quad \text{et} \quad \frac{d}{dt}T(t) = 0$$

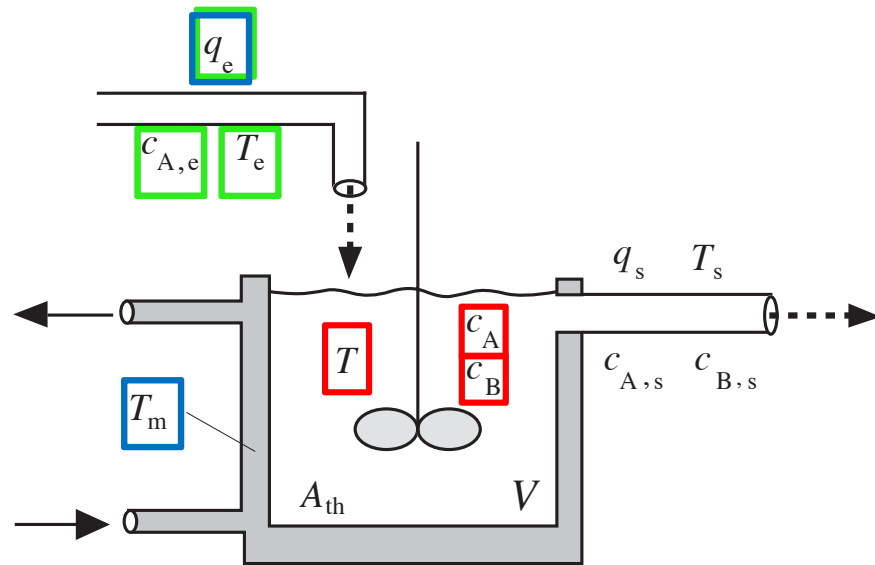
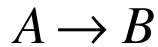
giving

$$\bar{w} := \bar{w}_e = \bar{w}_s$$

$$\bar{P} = \bar{w}c_p (\bar{T} - \bar{T}_e)$$

Continuously stirred reactor

Exothermic chemical reaction



Global mass balance

$$\frac{d}{dt}(\rho V) = \rho q_e(t) - \rho q_s(t) = 0$$

that is: $q(t) := q_e(t) = q_s(t)$

Assumptions

- First-order reaction

$$r(t) = k(t) c_A(t)$$

with

$$k(t) = k_0 \exp\left(\frac{-E_a}{RT(t)}\right)$$

(Arrhénius law)

- Well-mixed reactor:

$$T_s(t) = T(t)$$

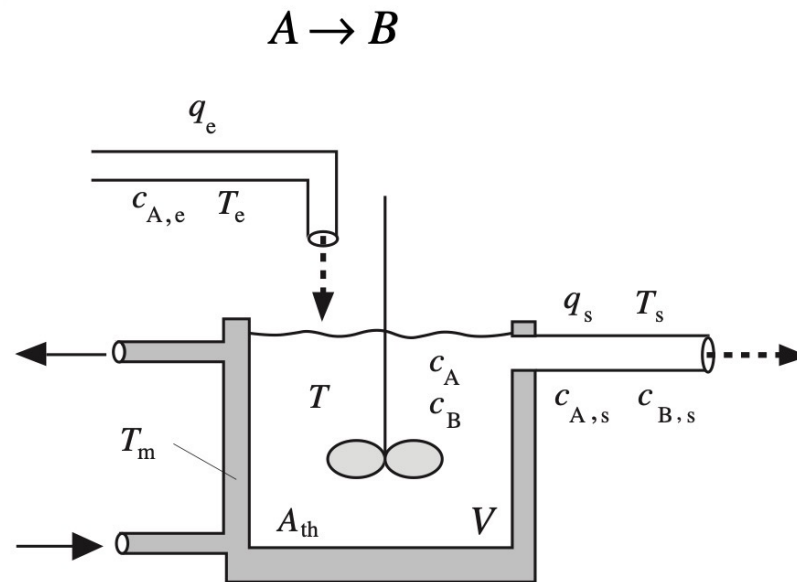
$$c_{A,s}(t) = c_A(t)$$

$$c_{B,s}(t) = c_B(t)$$

- Homogeneous cooling jacket
- Heat loss negligible

Continuously stirred reactor (cont'd)

Partial molar balance for component A



Continuously stirred reactor (cont'd)

Partial molar balance for component A

$$\left(\begin{array}{c} \text{variation in the number} \\ \text{of moles of A in the} \\ \text{reactor per unit of time} \end{array} \right) = \left(\begin{array}{c} \text{input molar} \\ \text{flow rate of A} \end{array} \right) - \left(\begin{array}{c} \text{output molar} \\ \text{flow rate of A} \end{array} \right) - \left(\begin{array}{c} \text{consumed number} \\ \text{of moles of A} \\ \text{per unit of time} \end{array} \right)$$

$$\frac{d}{dt}[Vc_A(t)] = q(t)c_{A,e}(t) - q(t)c_A(t) - Vr(t)$$

$$\frac{d}{dt}c_A(t) = \frac{q(t)}{V}[c_{A,e}(t) - c_A(t)] - k_0 \exp\left[\frac{-E_a}{RT(t)}\right]c_A(t)$$

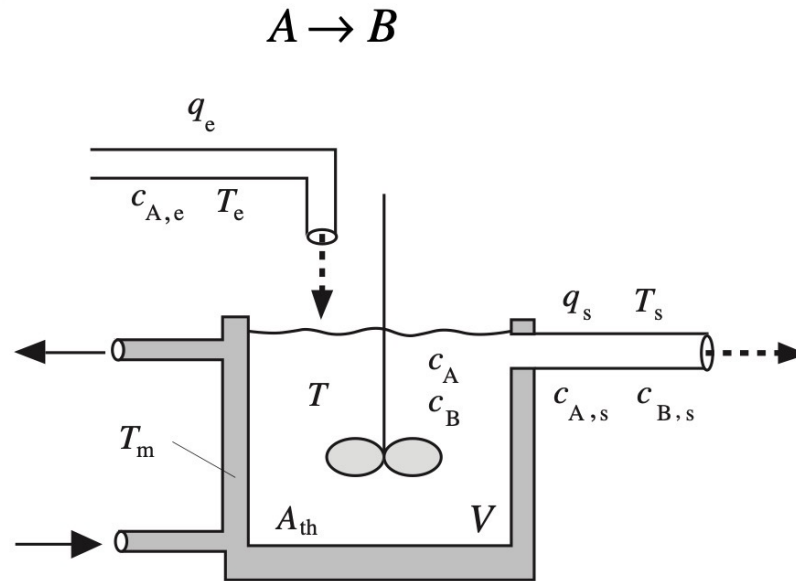
Partial molar balance for component B

$$\frac{d}{dt}[Vc_B(t)] = -q(t)c_B(t) + Vr(t)$$

$$\frac{d}{dt}c_B(t) = -\frac{q(t)}{V}c_B(t) + k_0 \exp\left[\frac{-E_a}{RT(t)}\right]c_A(t)$$

Continuously stirred reactor (cont'd)

Global energy balance



Continuously stirred reactor (cont'd)

Global energy balance

$$\left(\begin{array}{c} \text{Accumulated} \\ \text{power} \end{array} \right) = \left(\begin{array}{c} \text{Power exchanged} \\ \text{with the jacket} \end{array} \right) + \left(\begin{array}{c} \text{difference between} \\ \text{input and output} \\ \text{power} \end{array} \right) + \left(\begin{array}{c} \text{Power generated} \\ \text{by chemical reaction} \end{array} \right)$$

$$V\rho c_p \frac{d}{dt}T(t) = UA_{th} [T_m(t) - T(t)] + \rho q(t) c_p [T_e(t) - T(t)] + (-\Delta H) Vr(t)$$

$$\frac{d}{dt}T(t) = \frac{UA_{th}}{V\rho c_p} [T_m(t) - T(t)] + \frac{q(t)}{V} [T_e(t) - T(t)] + \frac{(-\Delta H)}{\rho c_p} k_0 \exp \left[\frac{-E_a}{RT(t)} \right] c_A(t)$$

Characteristic variables

- Constant parameters:
- Independent variables:
- Dependent variables (states):

$$V, \rho, c_p, U, A_{th}, k_0, R, E_a, (-\Delta H)$$

$$q(t), c_{A,e}(t), T_e(t), T_m(t)$$

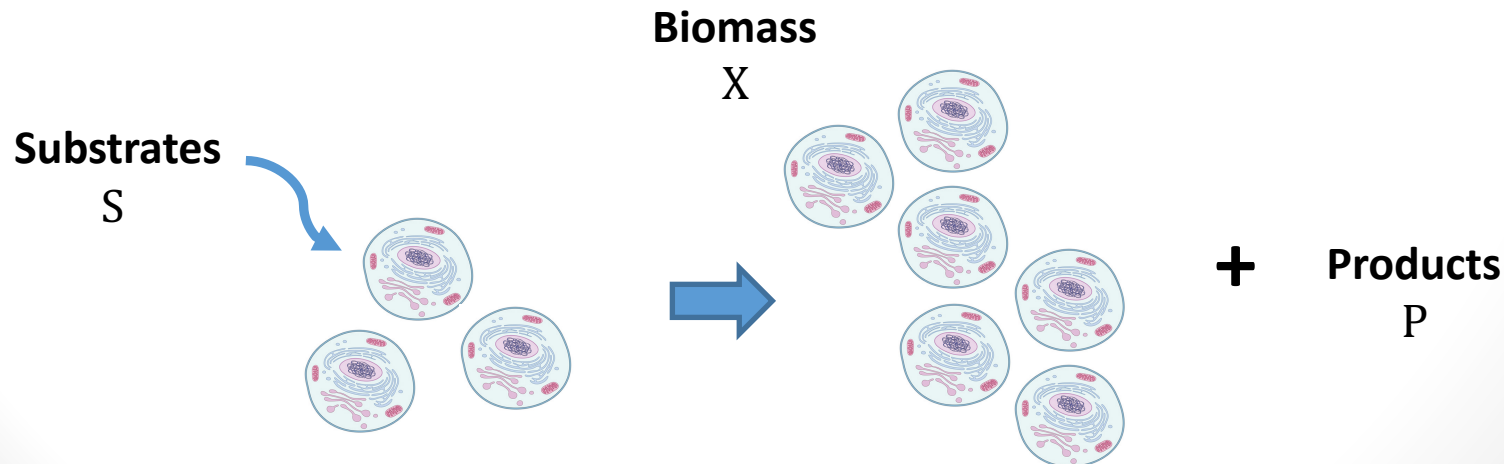
$$c_A(t), c_B(t), I(t)$$

Bioreactions

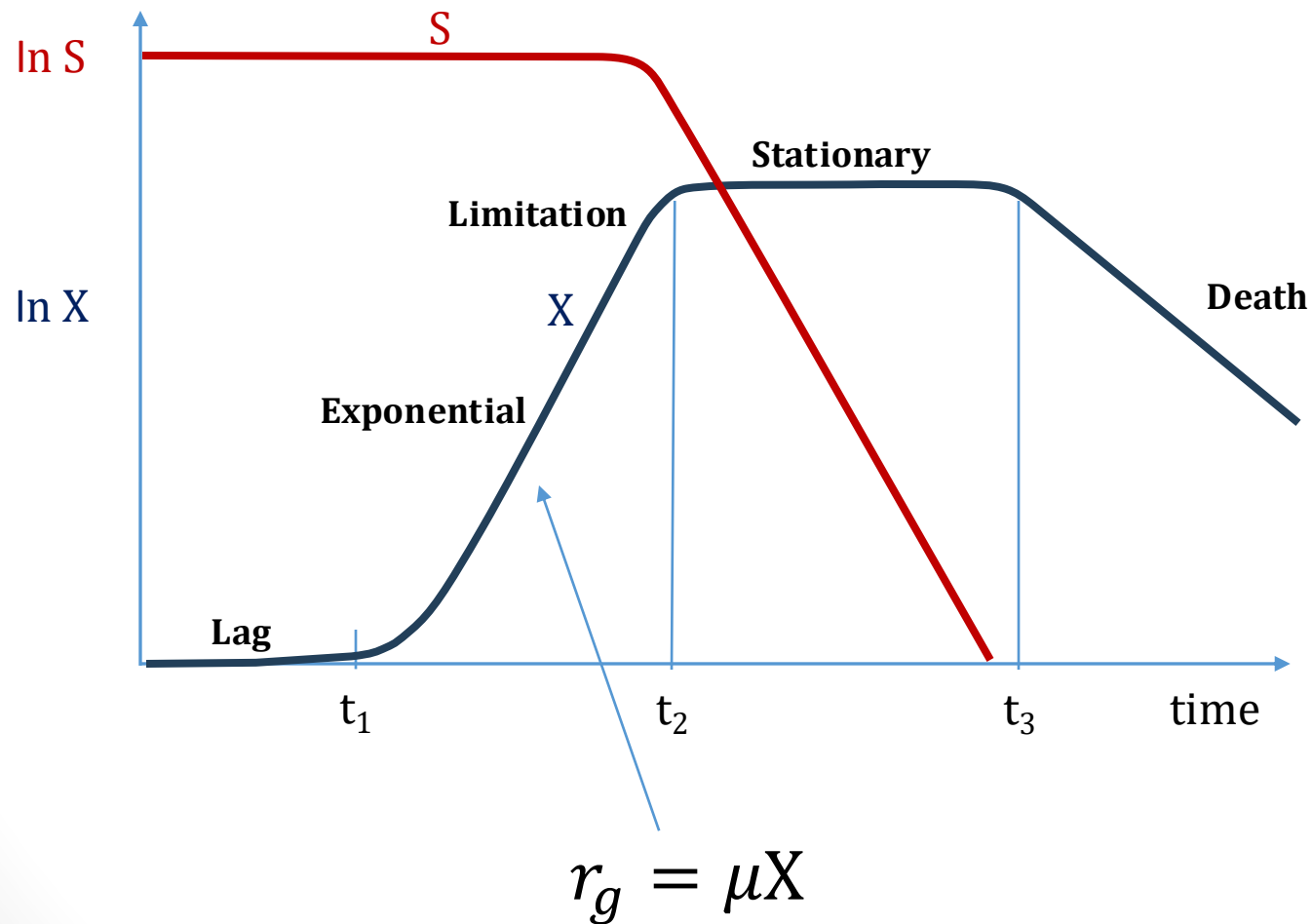
- Involve micro-organisms and enzyme catalysts.
- The basis for production of a wide variety of pharmaceuticals and healthcare and food products.
- Important industrial processes that involve bioreactions include fermentation and wastewater treatment.

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Basic growth kinetics in a batch



Growth kinetics

Monod Equation

$$r_g = \mu X$$

Specific Growth Rate

$$\mu = \mu_{max} \frac{S}{S + K_s}$$

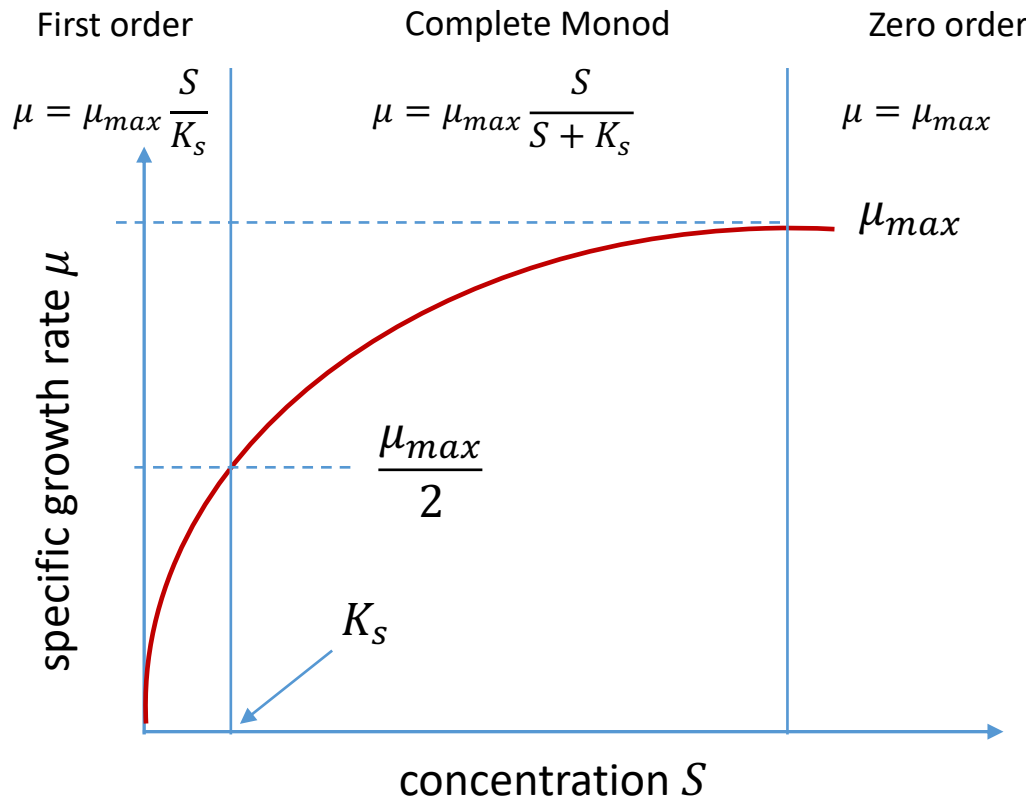
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Mass yield coefficients

The biomass yield on substrate

$$Y_{X/S} = \frac{r_g}{r_s}$$

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The product yield on substrate

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Mass yield coefficients

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The product yield on substrate

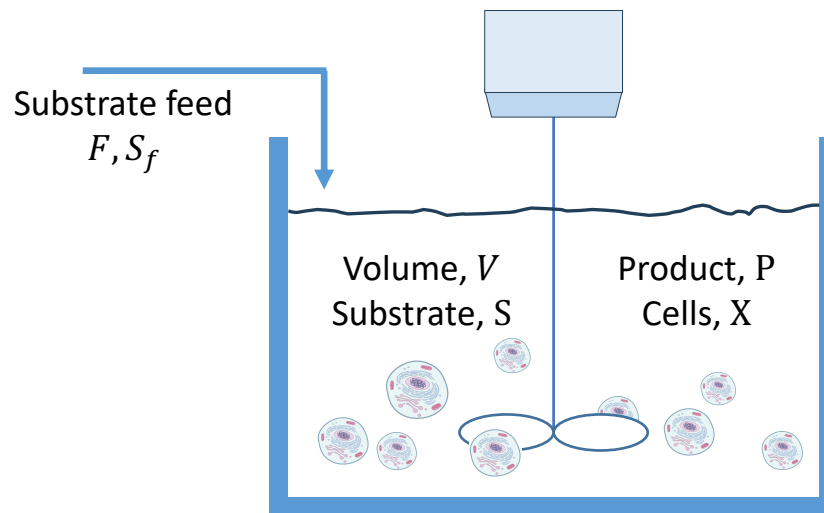
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The yield product-biomass

$$Y_{P/X} = \frac{r_p}{r_g} \quad Y_{P/X} = \frac{\text{mass of product formed}}{\text{mass of new cells formed}}$$

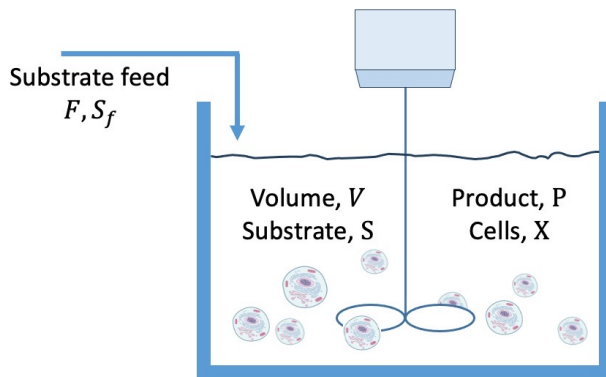
Fed-batch bioreactors

- Biological reactions typically performed in a batch or fed-batch reactor.
- Fed-batch reactors are widely used in the pharmaceutical and other process industries.



Fed-batch bioreactors

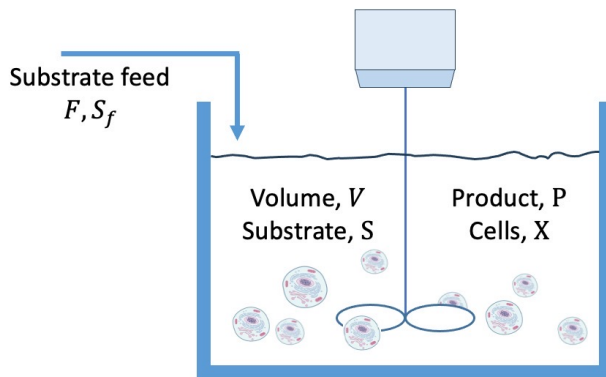
Modeling assumptions



1. The exponential cell growth stage
2. Perfectly mixed broth (cells + liquid), homogeneous
3. Isothermal reactor operation
4. The liquid density is constant
5. The cell growth rate r_g given by the Monod equation

Fed-batch bioreactors

Modeling assumptions



1. The exponential cell growth stage
2. Perfectly mixed broth (cells + liquid), homogeneous
3. Isothermal reactor operation
4. The liquid density is constant
5. The cell growth rate r_g given by the Monod equation
6. The rate of product formation per unit volume r_p can be expressed as

$$r_p = Y_{P/X} r_g$$

$$Y_{P/X} = \frac{\text{mass of product formed}}{\text{mass of new cells formed}}$$

Fed-batch bioreactors

Individual Component Balances

Cells:

$$\frac{d(XV)}{dt} = Vr_g$$

$$r_g = \mu_{max} \frac{S}{S + K_s} X$$


Fed-batch bioreactors

Individual Component Balances

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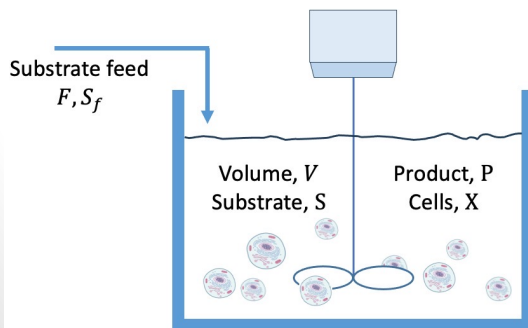
$$r_g = \mu_{max} \frac{S}{S + K_s} X$$

Product:

$$\frac{d(PV)}{dt} = Vr_p$$

Substrate:

$$\frac{d(SV)}{dt} = FS_f - \frac{1}{Y_{X/S}} Vr_g$$



Note: in some books Yield coefficients are defined as

$$Y_{X/S}^* = \frac{\text{mass of new cells formed}}{\text{mass of substrate consumed to form cells}}$$

$$Y_{P/S}^* = \frac{\text{mass of product formed}}{\text{mass of substrate consumed to form product}}$$

$$\frac{d(SV)}{dt} = FS_f - \frac{1}{Y_{X/S}^*} Vr_g - \frac{1}{Y_{P/S}^*} Vr_p$$

Fed-batch bioreactors

Individual Component Balances

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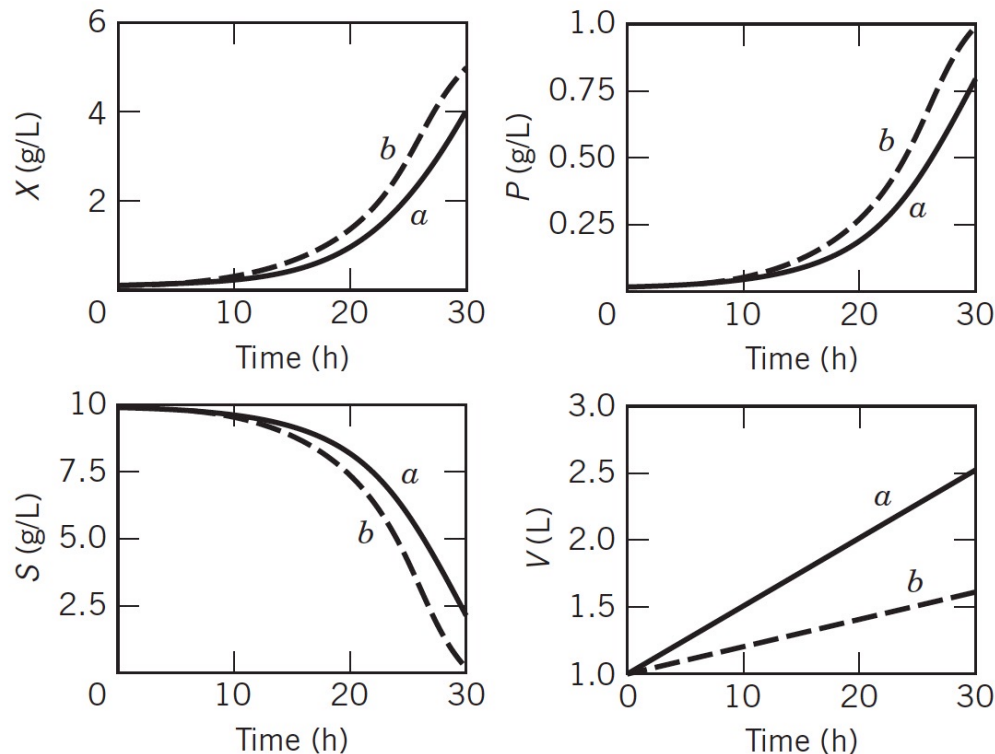
Overall Mass Balance

Mass:

$$\frac{dV}{dt} = F$$

Fed-batch bioreactor simulation*

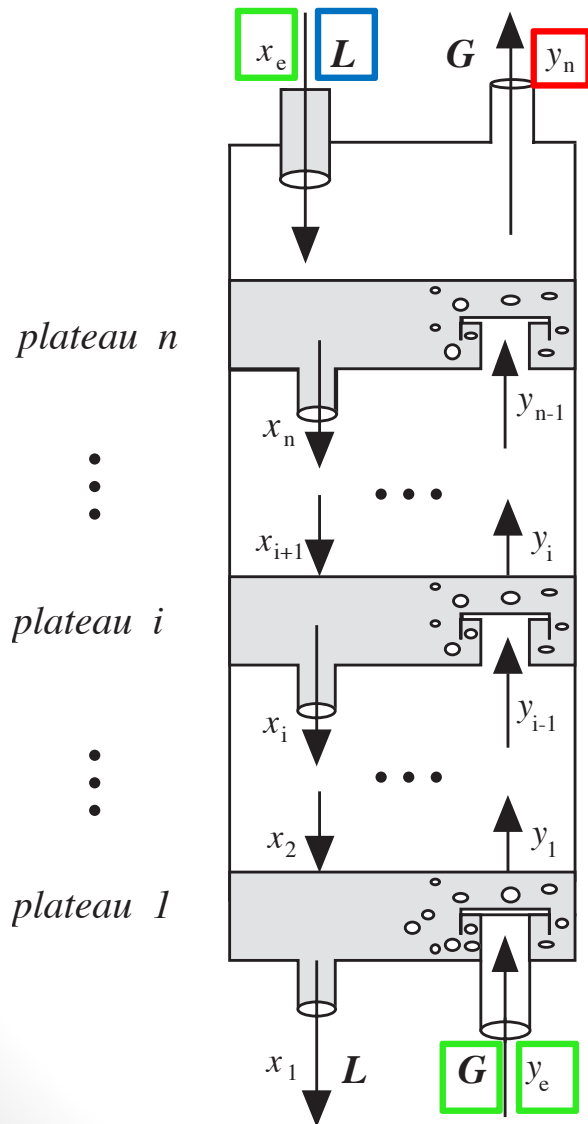
a (solid): $F = 0.05$ L/h; b (dashed): $F = 0.02$ L/h.



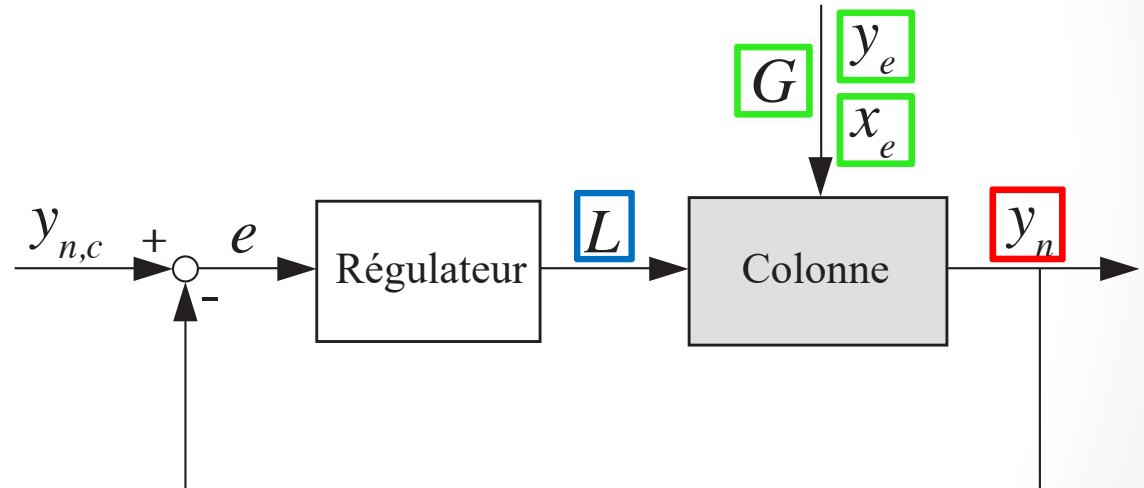
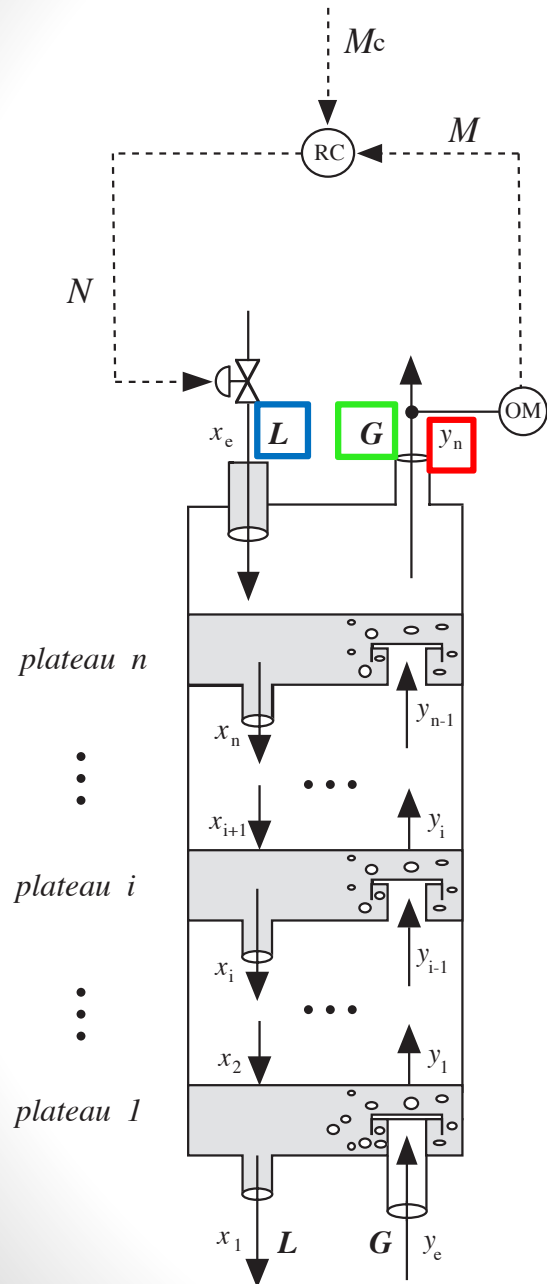
Model Parameters			Simulation Conditions		
$\mu_{\max}$	0.20	h^{-1}	S_f	10.0	g/L
K_S	1.0	g/L	$X(0)$	0.05	g/L
$Y_{X/S}$	0.5	g/g	$S(0)$	10.0	g/L
$Y_{P/X}$	0.2	g/g	$P(0)$	0.0	g/L
			$V(0)$	1.0	L

* from Process Dynamic and control by Seborg et al.

Absorption column



Control of the absorption column



State-space representation of systems



$$\begin{aligned}\dot{y}(t) &= -3y(t) + 2u(t)z(t) & y(0) \\ \dot{z}(t) &= 2y(t) - z(t) & z(0)\end{aligned}$$

State-space representation of systems



$$\begin{aligned}\dot{y}(t) &= -3y(t) + 2u(t)z(t) & y(0) \\ \dot{z}(t) &= 2y(t) - z(t) & z(0)\end{aligned}$$

Behavior of $y(t)$ and $z(t)$ for $t \geq 0$ depends on $u(t \geq 0)$ **but also** on $y(0)$ and $z(0)$!

State-space representation of systems



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Behavior of $y(t)$ and $z(t)$ for $t \geq 0$ depends on $u(t \geq 0)$ **but also** on $y(0)$ and $z(0)$!

The minimal required information to unambiguously determine future behavior of a dynamic deterministic system for $t \geq t_0$ is:

- the **state** of system at $t = t_0$
- the knowledge of the future inputs, i.e., $u(t \geq t_0)$

State-space models

$$\begin{aligned} \dot{x}_1(t) &= f_1[x_1(t), \dots, x_n(t), u_1(t), \dots, u_p(t), t] & x_1(t_0) &= x_{1,0} \\ \dot{x}_2(t) &= f_2[x_1(t), \dots, x_n(t), u_1(t), \dots, u_p(t), t] & x_2(t_0) &= x_{2,0} \\ &\vdots & & \\ \dot{x}_n(t) &= f_n[x_1(t), \dots, x_n(t), u_1(t), \dots, u_p(t), t] & x_n(t_0) &= x_{n,0} \end{aligned}$$

$$\begin{aligned} y_1(t) &= g_1[x_1(t), \dots, x_n(t), u_1(t), \dots, u_p(t), t] \\ y_2(t) &= g_2[x_1(t), \dots, x_n(t), u_1(t), \dots, u_p(t), t] \\ &\vdots \\ y_q(t) &= g_q[x_1(t), \dots, x_n(t), u_1(t), \dots, u_p(t), t] \end{aligned}$$

State-space models

$$x(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix}$$

Vector of states

$$u(t) = \begin{bmatrix} u_1(t) \\ u_2(t) \\ \vdots \\ u_p(t) \end{bmatrix}$$

Vector of inputs

$$y(t) = \begin{bmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_q(t) \end{bmatrix}$$

Vector of outputs

State equation

$$\dot{x}(t) = f[x(t), u(t), t] \quad x(t_0) = x_0$$

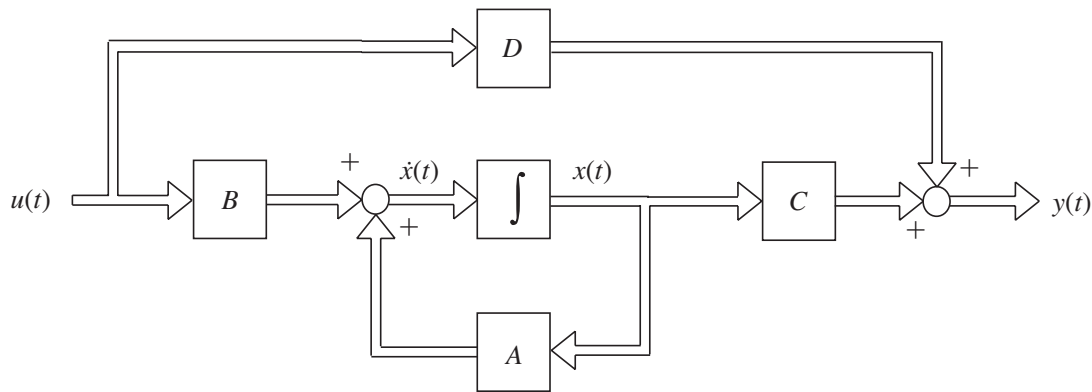
Output equation

$$y(t) = g[x(t), u(t), t]$$

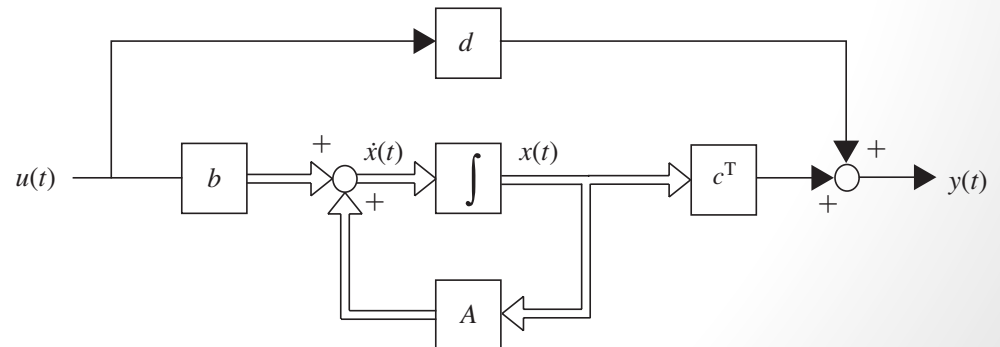
Linear, stationary, state-space models

$$\dot{x}(t) = Ax(t) + Bu(t) \quad x(0) = x_0$$

$$y(t) = Cx(t) + Du(t)$$

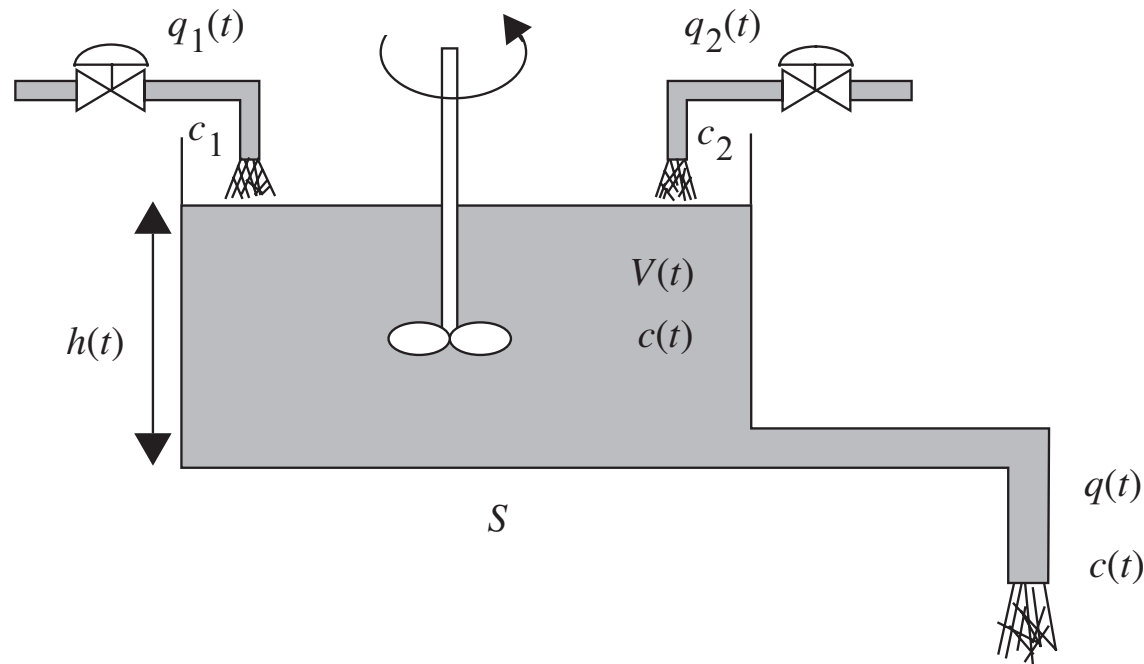


Multivariable



Monovariable

Mixing tank example



$$q(t) = k\sqrt{h(t)}$$

$$\rho S \frac{d}{dt} h(t) = \rho q_1(t) + \rho q_2(t) - \rho q(t) \quad \left[\frac{\text{kg}}{\text{s}} \right]$$

$$\rho S \frac{d}{dt} [c(t)h(t)] = \rho c_1 q_1(t) + \rho c_2 q_2(t) - \rho c(t) q(t) \quad \left[\frac{\text{kgP}}{\text{s}} \right]$$

Mixing tank state-space model

$$x_1(t) = h(t)$$

$$x_2(t) = c(t)$$

states

$$u_1(t) = q_1(t)$$

$$u_2(t) = q_2(t)$$

inputs

$$y_1(t) = h(t)$$

$$y_2(t) = \frac{c(t)}{c_1}$$

outputs

$$\dot{x}_1(t) = \frac{1}{S} [u_1(t) + u_2(t) - k\sqrt{x_1(t)}]$$

$$x_1(0) = h(0)$$

→

$$\dot{x}_2(t) = \frac{1}{Sx_1(t)} \{ [c_1 - x_2(t)]u_1(t) + [c_2 - x_2(t)]u_2(t) \}$$

$$x_2(0) = c(0)$$

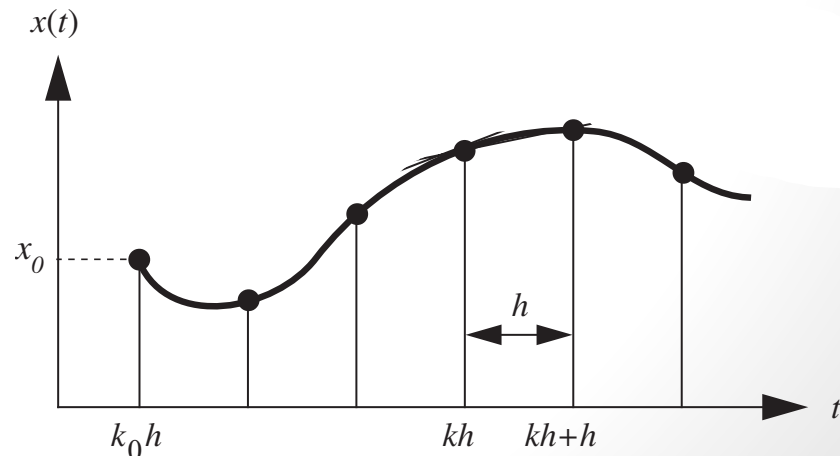
Once having a nonlinear model – what to do?

Simulation: Determine the state $x(t)$ and the output $y(t)$ for $t \geq t_0$ starting from $x(t_0)$ and using the knowledge of $u(t)$ for $t \geq t_0$.

For $t = kh$

$$\dot{x}(kh) = f[x(kh), u(kh)] \quad x(k_0h) = x_0$$

$$y(kh) = g[x(kh), u(kh)]$$



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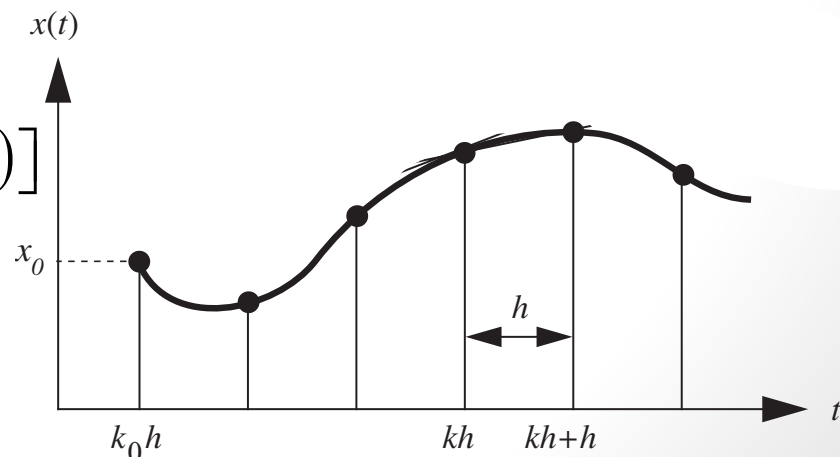
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Euler

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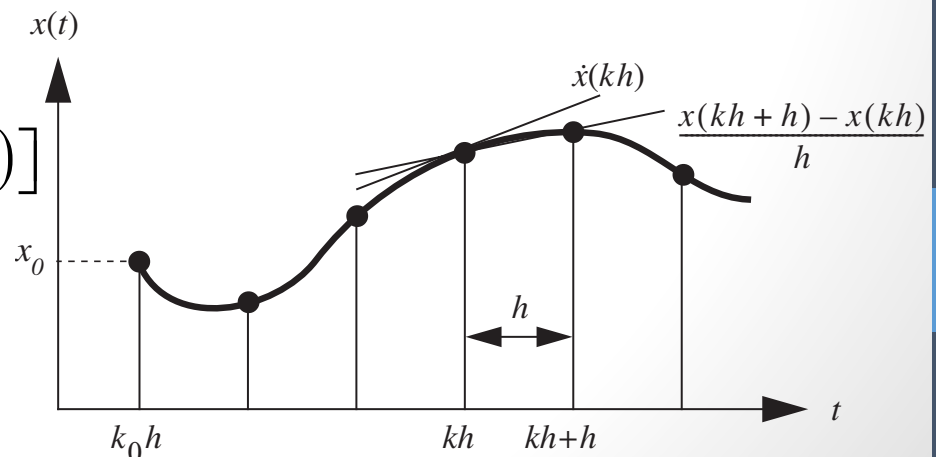
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Exact (feedback) linearization

Nonlinear model

$$\dot{x}_1(t) = \frac{1}{S} \left[u_1(t) + u_2(t) - k\sqrt{x_1(t)} \right]$$


$$A\dot{h}(t) = u(t) - k\sqrt{h(t)}$$

Exact (feedback) linearization

Nonlinear model

$$A\dot{h}(t) = u(t) - k\sqrt{h(t)}$$

If we could measure $h(t)$, then we could define a new input that satisfies

$$u(t) - k\sqrt{h(t)} = Av(t)$$

Exact (feedback) linearization

Nonlinear model

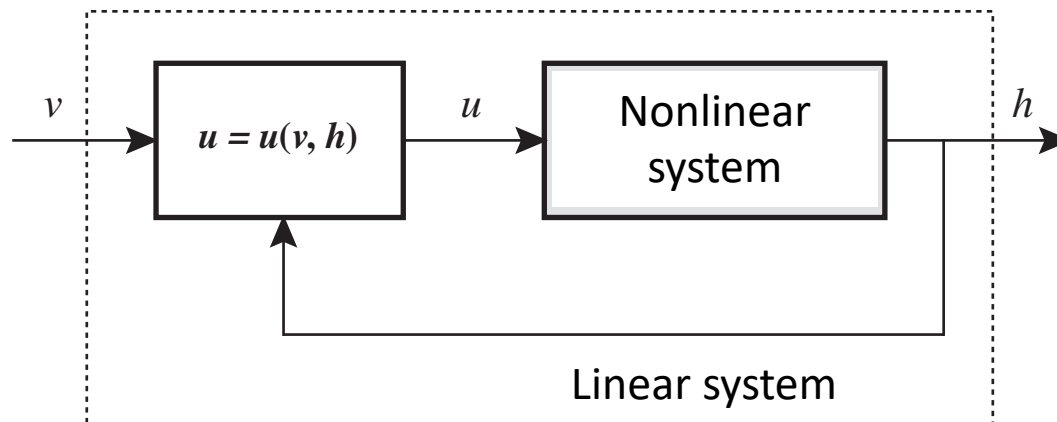
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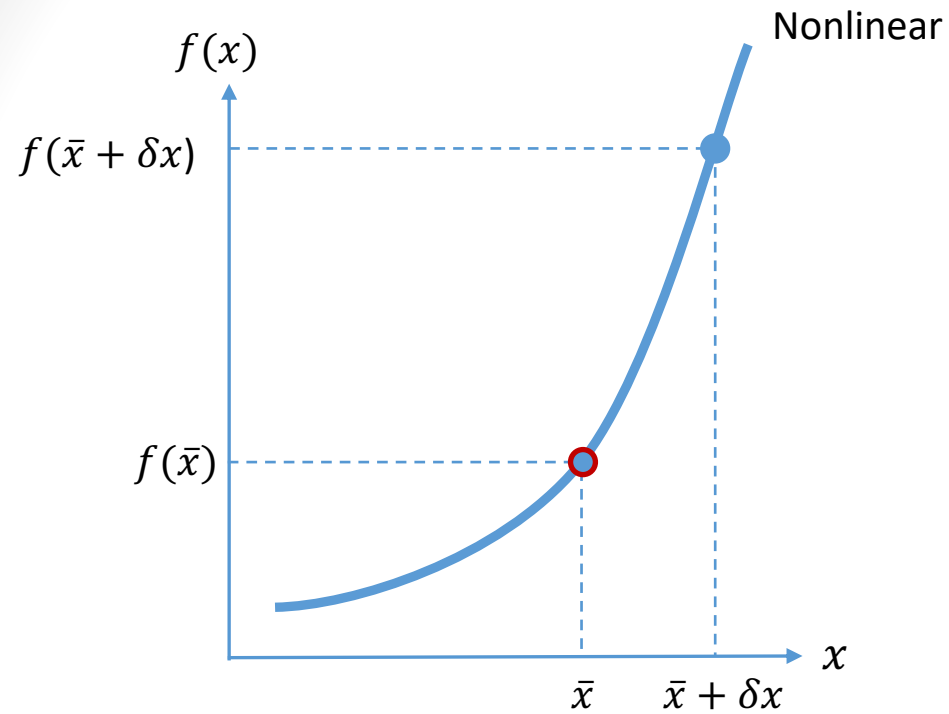
$$u(t) - k\sqrt{h(t)} = Av(t)$$

And our system becomes linear:

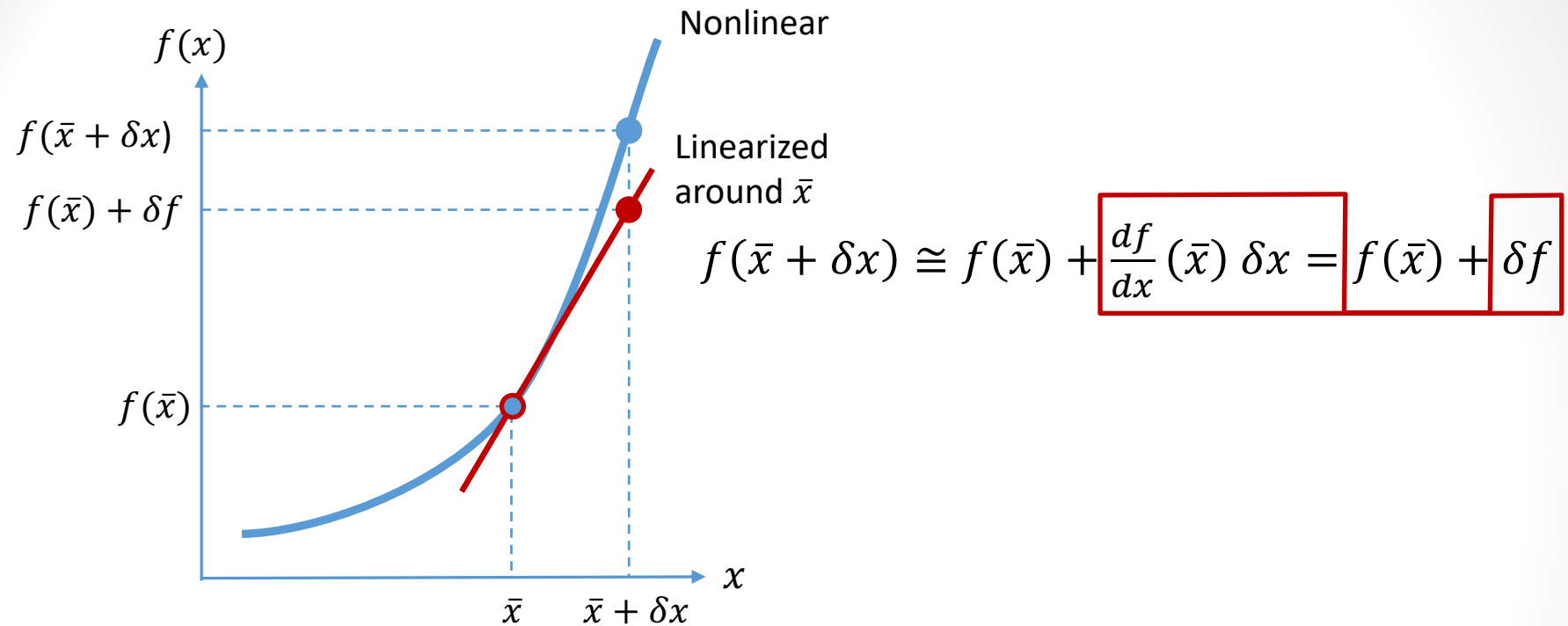
$$\dot{h}(t) = v(t)$$



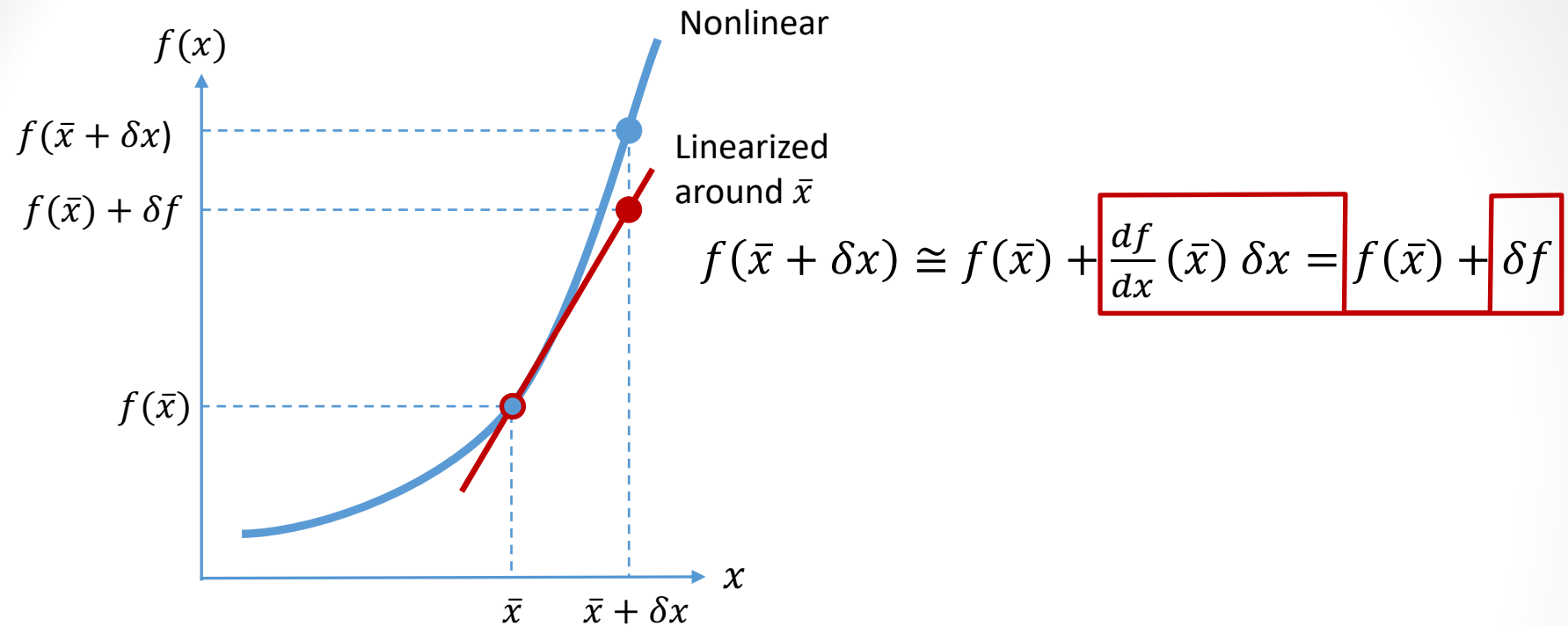
Approximate linearization



Approximate linearization



Approximate linearization



Depending on the control problem (tracking or regulation), we can perform linearization around:

Nominal trajectories

$$\dot{\bar{x}} = f[\bar{x}(t), \bar{u}(t), t] \quad \bar{x}(0) = x_0$$

$$\bar{y}(t) = g[\bar{x}(t), \bar{u}(t), t]$$

Stationary point

$$0 = f[\bar{x}, \bar{u}]$$

$$\bar{y} = g[\bar{x}, \bar{u}]$$

Linearization procedure

$$\dot{x}(t) = f[x(t), u(t)] \quad x(0) = x_0$$

$$y(t) = g[x(t), u(t)]$$

Equilibrium point

$$0 = f[\bar{x}, \bar{u}]$$

$$\bar{y} = g[\bar{x}, \bar{u}]$$

Linearization procedure

$$\begin{aligned}\dot{x}(t) &= f[x(t), u(t)] & x(0) &= x_0 \\ y(t) &= g[x(t), u(t)]\end{aligned}$$

Equilibrium point

$$\begin{aligned}0 &= f[\bar{x}, \bar{u}] \\ \bar{y} &= g[\bar{x}, \bar{u}]\end{aligned}$$

Taylor series

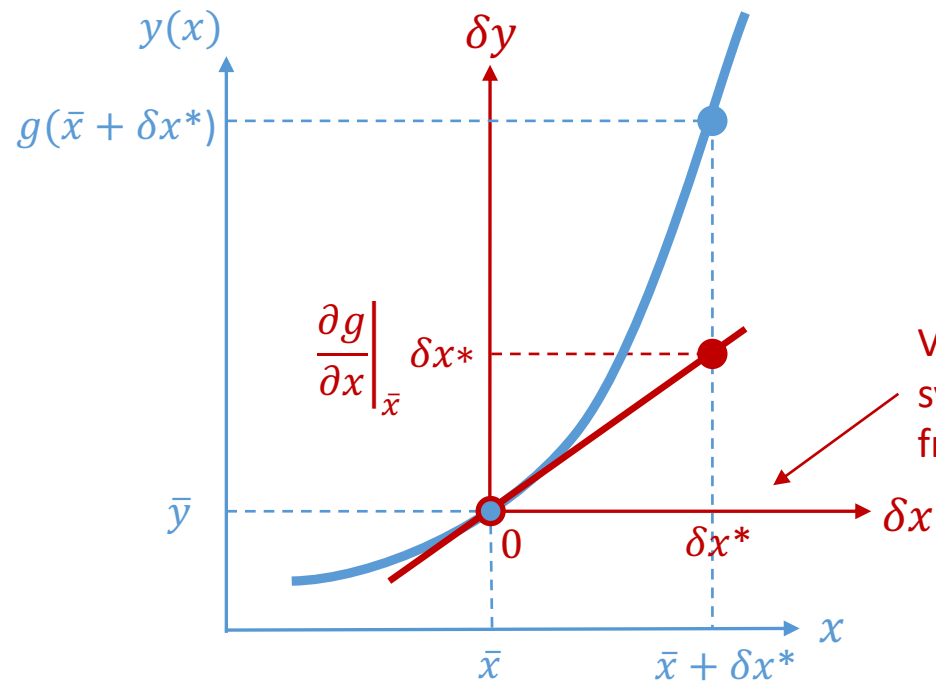
$$\dot{x} = f[\bar{x}, \bar{u}] + \left. \frac{\partial f}{\partial x} \right|_{\bar{u}, \bar{x}} (x - \bar{x}) + \left. \frac{\partial f}{\partial u} \right|_{\bar{u}, \bar{x}} (u - \bar{u}) + \text{Higher-order terms}$$

$$y = g[\bar{x}, \bar{u}] + \left. \frac{\partial g}{\partial x} \right|_{\bar{u}, \bar{x}} (x - \bar{x}) + \left. \frac{\partial g}{\partial u} \right|_{\bar{u}, \bar{x}} (u - \bar{u}) + \text{Higher-order terms}$$

Linearization procedure (cont'd)

$$\dot{x} = f[\bar{x}, \bar{u}] + \left. \frac{\partial f}{\partial x} \right|_{\bar{u}, \bar{x}} (x - \bar{x}) + \left. \frac{\partial f}{\partial u} \right|_{\bar{u}, \bar{x}} (u - \bar{u})$$

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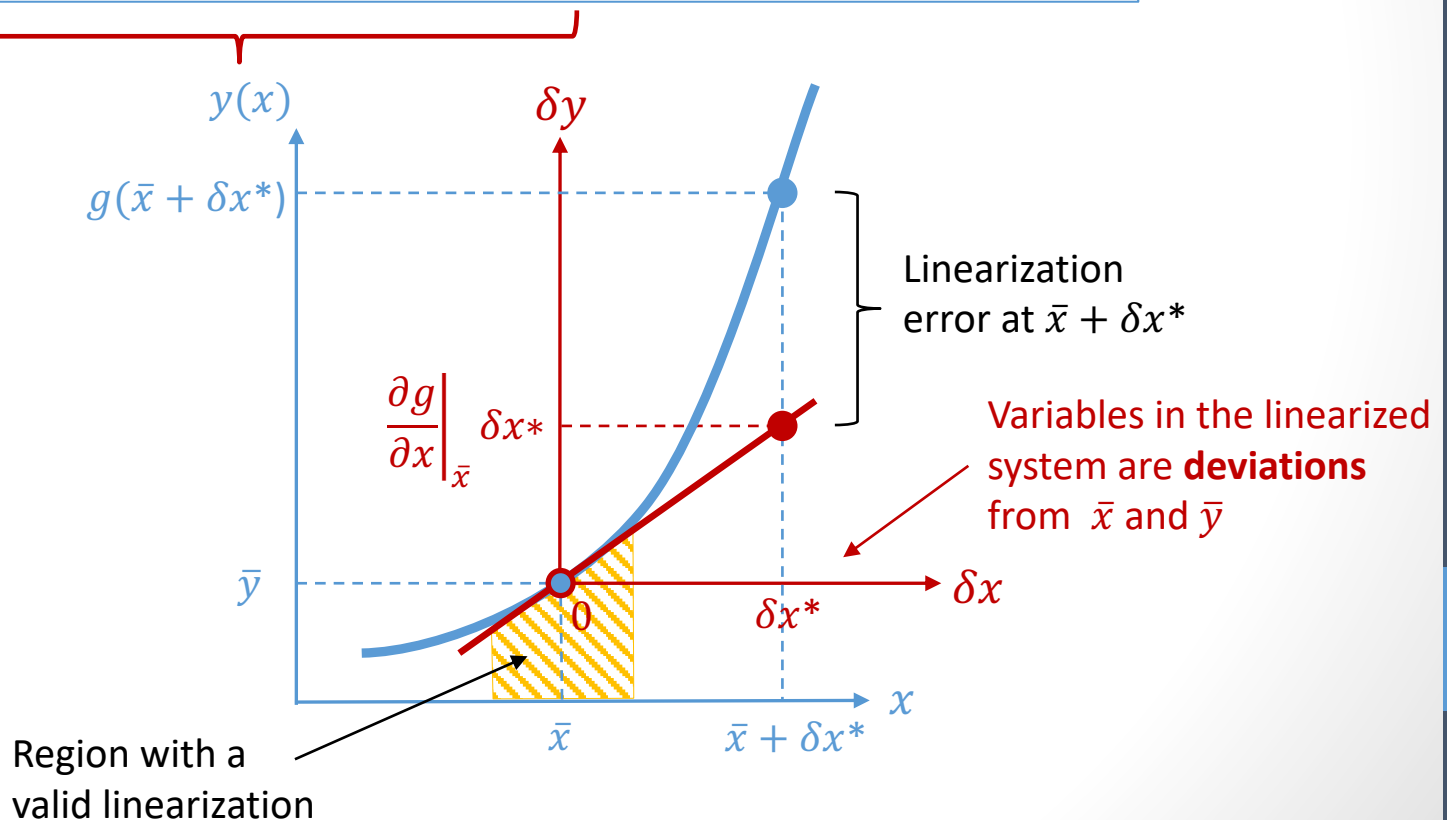


Variables in the linearized system are deviations from $\bar{x}$ and $\bar{y}$

Linearization procedure (cont'd)

$$\dot{x} = f[\bar{x}, \bar{u}] + \left. \frac{\partial f}{\partial x} \right|_{\bar{u}, \bar{x}} (x - \bar{x}) + \left. \frac{\partial f}{\partial u} \right|_{\bar{u}, \bar{x}} (u - \bar{u})$$

$$y = g[\bar{x}, \bar{u}] + \left. \frac{\partial g}{\partial x} \right|_{\bar{u}, \bar{x}} (x - \bar{x}) + \left. \frac{\partial g}{\partial u} \right|_{\bar{u}, \bar{x}} (u - \bar{u})$$



Linearization procedure (cont'd)

Deviations

$$\delta x(t) := x(t) - \bar{x}$$

$$\delta u(t) := u(t) - \bar{u}$$

$$\delta y(t) := y(t) - \bar{y}$$


$$\begin{aligned}\dot{x} &= f[\bar{x}, \bar{u}] + \left. \frac{\partial f}{\partial x} \right|_{\bar{u}, \bar{x}} (x - \bar{x}) + \left. \frac{\partial f}{\partial u} \right|_{\bar{u}, \bar{x}} (u - \bar{u}) + \text{tos} \\ y &= g[\bar{x}, \bar{u}] + \left. \frac{\partial g}{\partial x} \right|_{\bar{u}, \bar{x}} (x - \bar{x}) + \left. \frac{\partial g}{\partial u} \right|_{\bar{u}, \bar{x}} (u - \bar{u}) + \text{tos}\end{aligned}$$

Linearization procedure (cont'd)

Deviations

$$\delta x(t) := x(t) - \bar{x}$$

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Linear approximation

$$\delta \dot{x} = \left. \frac{\partial f}{\partial x} \right|_{\bar{u}, \bar{x}} \delta x + \left. \frac{\partial f}{\partial u} \right|_{\bar{u}, \bar{x}} \delta u$$

$$\delta x(0) = x_0 - \bar{x}$$

$$\delta y = \left. \frac{\partial g}{\partial x} \right|_{\bar{u}, \bar{x}} \delta x + \left. \frac{\partial g}{\partial u} \right|_{\bar{u}, \bar{x}} \delta u$$

$$\delta \dot{x} = A\delta x + B\delta u \quad \delta x(0) = x_0 - \bar{x}$$

$$\delta y = C\delta x + D\delta u$$

Toy example: method 1

$$\dot{x} = -2x + 0,5(x+1)u \quad x(0) = 1$$

Equilibrium point

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$$0 = -2\bar{x} + 0,5(\bar{x}+1)\bar{u}$$

For $\bar{u} = 2$, we obtain $\bar{x} = 1$

Linear approximation of the nonlinearity

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Linear approximation of the nonlinearity

$$xu \simeq \bar{x}\bar{u} + \bar{u}(x - \bar{x}) + \bar{x}(u - \bar{u})$$

Deviations

$$\delta x := x - \bar{x} \quad \delta u := u - \bar{u}$$

$$xu \cong \bar{x}\bar{u} + \bar{u}\delta x + \bar{x}\delta u$$

Linear approximation of the overall system

$$\dot{x} = \left[-2x + 0,5(\bar{x}\bar{u} + \bar{u}\delta x + \bar{x}\delta u) + 0,5u \right]$$

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$$\delta x := x - \bar{x} \quad \delta u := u - \bar{u}$$

Linear approximation of the overall system

$$\begin{aligned} \dot{x} &= \left[-2x + 0,5(\bar{x}\bar{u} + \bar{u}\delta x + \bar{x}\delta u) + 0,5u \right] - \left[-2\bar{x} + 0,5\bar{x}\bar{u} + 0,5\bar{u} \right] \\ &= -2\delta x + 0,5(\bar{u}\delta x + \bar{x}\delta u) + 0,5\delta u \end{aligned}$$

$$\delta \dot{x} = \dot{x}$$

$$\delta \dot{x} = -\delta x + \delta u \quad \delta x(0) = 0$$

Reminder: state-space matrices

Linear approximation

$$\delta \dot{x} = \left. \frac{\partial f}{\partial x} \right|_{\bar{u}, \bar{x}} \delta x + \left. \frac{\partial f}{\partial u} \right|_{\bar{u}, \bar{x}} \delta u \quad \delta x(0) = x_0 - \bar{x}$$

$$\delta y = \left. \frac{\partial g}{\partial x} \right|_{\bar{u}, \bar{x}} \delta x + \left. \frac{\partial g}{\partial u} \right|_{\bar{u}, \bar{x}} \delta u$$

$$\delta \dot{x} = A \delta x + B \delta u \quad \delta x(0) = x_0 - \bar{x}$$

$$\delta y = C \delta x + D \delta u$$

State-space representation of linear system

$$\dot{x}(t) = Ax(t) + Bu(t) \quad x(0) = x_0$$

$$y(t) = Cx(t) + Du(t)$$

State-space matrices

$$A: = \frac{\partial f}{\partial x} \Big|_{\bar{u}, \bar{x}} = \left[\begin{array}{cccc} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{array} \right]_{\bar{u}, \bar{x}} \quad (n \times n)$$

$$B: = \frac{\partial f}{\partial u} \Big|_{\bar{u}, \bar{x}} = \left[\begin{array}{cccc} \frac{\partial f_1}{\partial u_1} & \frac{\partial f_1}{\partial u_2} & \dots & \frac{\partial f_1}{\partial u_p} \\ \frac{\partial f_2}{\partial u_1} & \frac{\partial f_2}{\partial u_2} & \dots & \frac{\partial f_2}{\partial u_p} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_n}{\partial u_1} & \frac{\partial f_n}{\partial u_2} & \dots & \frac{\partial f_n}{\partial u_p} \end{array} \right]_{\bar{u}, \bar{x}} \quad (n \times p)$$

$$C: = \frac{\partial g}{\partial x} \Big|_{\bar{u}, \bar{x}} = \left[\begin{array}{cccc} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} & \dots & \frac{\partial g_1}{\partial x_n} \\ \frac{\partial g_2}{\partial x_1} & \frac{\partial g_2}{\partial x_2} & \dots & \frac{\partial g_2}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial g_q}{\partial x_1} & \frac{\partial g_q}{\partial x_2} & \dots & \frac{\partial g_q}{\partial x_n} \end{array} \right]_{\bar{u}, \bar{x}} \quad (q \times n)$$

$$D: = \frac{\partial g}{\partial u} \Big|_{\bar{u}, \bar{x}} = \left[\begin{array}{cccc} \frac{\partial g_1}{\partial u_1} & \frac{\partial g_1}{\partial u_2} & \dots & \frac{\partial g_1}{\partial u_p} \\ \frac{\partial g_2}{\partial u_1} & \frac{\partial g_2}{\partial u_2} & \dots & \frac{\partial g_2}{\partial u_p} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial g_q}{\partial u_1} & \frac{\partial g_q}{\partial u_2} & \dots & \frac{\partial g_q}{\partial u_p} \end{array} \right]_{\bar{u}, \bar{x}} \quad (q \times p)$$

Toy example: method 2

$$\dot{x} = -2x + 0,5(x+1)u \quad x(0) = 1$$

Monovariate system: $n = p = 1$

Toy example: method 2

$$\dot{x} = -2x + 0,5(x+1)u \quad x(0) = 1$$

Monovariate system: $n = p = 1$

$$\dot{x} = f(x, u) \rightarrow f(x, u) = -2x + 0.5xu + 0.5u$$

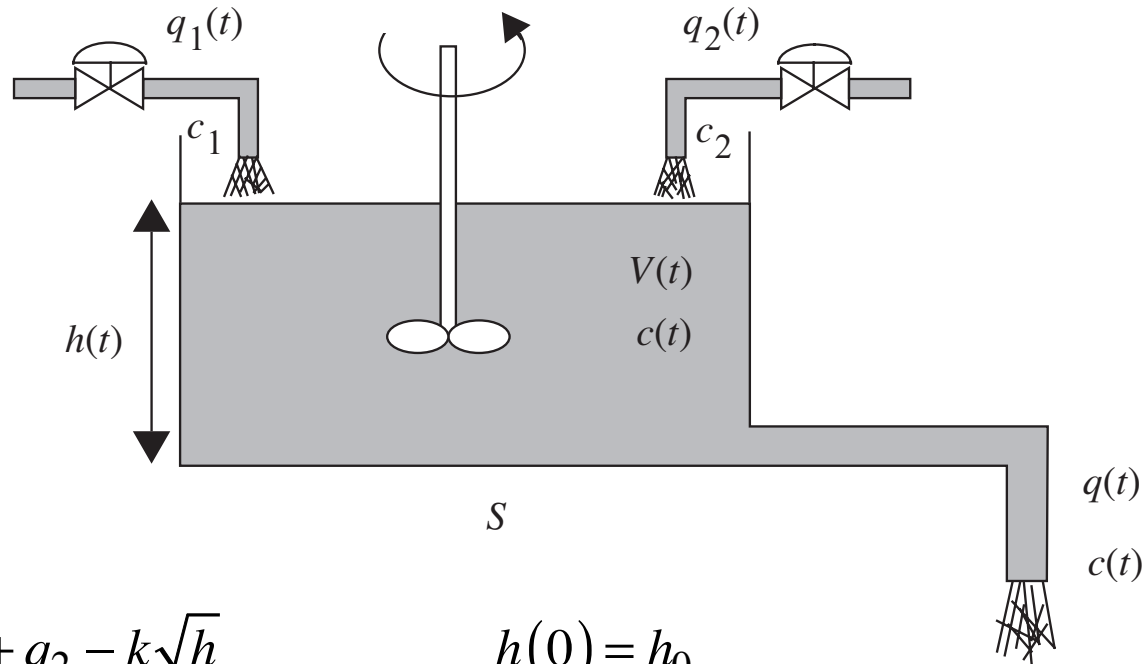
$$A = \left. \frac{\partial f(x, u)}{\partial x} \right|_{\bar{x}, \bar{u}} = \left. \frac{\partial (-2x + 0.5xu + 0.5u)}{\partial x} \right|_{\bar{x}, \bar{u}} = -2 + 0.5\bar{u} = -1$$

$$B = \left. \frac{\partial f(x, u)}{\partial u} \right|_{\bar{x}, \bar{u}} = 0.5\bar{x} + 0.5 = 1$$

$$\delta \dot{x} = A\delta x + B\delta u = -\delta x + \delta u$$

$$\delta x(0) = x(0) - \bar{x} = 0$$

Mixing tank



$$S\dot{h} = q_1 + q_2 - k\sqrt{h} \quad h(0) = h_0$$

$$\dot{c} = \frac{q_1}{Sh}(c_1 - c) + \frac{q_2}{Sh}(c_2 - c) \quad c(0) = c_0$$

$$y_1 = h$$

$$y_2 = \frac{c}{c_1}$$

Multivariable system $n = p = q = 2$

Mixing tank (cont'd)

Operating point

$$(\bar{q}_1, \bar{q}_2, \bar{h}, \bar{c})$$

$$\bar{q}_1, \bar{q}_2 \rightarrow \bar{u}_1, \bar{u}_2$$

$$\bar{h}, \bar{c} \rightarrow \bar{x}_1, \bar{x}_2$$

Multivariable system $n = p = q = 2$

$$A = \left. \frac{\partial f(x,u)}{\partial x} \right|_{\bar{x}, \bar{u}} = \left[\begin{array}{cc} \frac{\partial f_1(x,u)}{\partial x_1} & \frac{\partial f_1(x,u)}{\partial x_2} \\ \frac{\partial f_2(x,u)}{\partial x_1} & \frac{\partial f_2(x,u)}{\partial x_2} \end{array} \right]_{\bar{x}, \bar{u}}$$

$$\dot{h} = (q_1 + q_2 - k\sqrt{h})/S \rightarrow f_1$$

$$\dot{c} = \frac{q_1}{Sh} (c_1 - c) + \frac{q_2}{Sh} (c_2 - c) \rightarrow f_2$$

Mixing tank (cont'd)

Operating point

$$(\bar{q}_1, \bar{q}_2, \bar{h}, \bar{c})$$

$$\bar{q}_1, \bar{q}_2 \rightarrow \bar{u}_1, \bar{u}_2$$

$$\bar{h}, \bar{c} \rightarrow \bar{x}_1, \bar{x}_2$$

$$A = \left. \frac{\partial f(x,u)}{\partial x} \right|_{\bar{x}, \bar{u}} = \begin{bmatrix} \frac{\partial f_1(x,u)}{\partial x_1} & \frac{\partial f_1(x,u)}{\partial x_2} \\ \frac{\partial f_2(x,u)}{\partial x_1} & \frac{\partial f_2(x,u)}{\partial x_2} \end{bmatrix} \bigg|_{\bar{x}, \bar{u}}$$

$$\dot{h} = (q_1 + q_2 - k\sqrt{h})/S \rightarrow f_1$$

$$\dot{c} = \frac{q_1}{Sh} (c_1 - c) + \frac{q_2}{Sh} (c_2 - c) \rightarrow f_2$$

Multivariable system $n = p = q = 2$

$$a_{11} = \left. \frac{\partial f_1}{\partial h} \right|_{eq.p.} = -\frac{k}{2S\sqrt{\bar{h}}} \quad a_{12} = \left. \frac{\partial f_1}{\partial c} \right|_{eq.p.} = 0$$

$$a_{21} = \left. \frac{\partial f_2}{\partial h} \right|_{eq.p.} = -\frac{\bar{q}_1}{S\bar{h}^2} (c_1 - \bar{c}) - \frac{\bar{q}_2}{S\bar{h}^2} (c_2 - \bar{c}) = \frac{1}{S\bar{h}^2} [\bar{c}(\bar{q}_1 + \bar{q}_2) - c_1\bar{q}_1 - c_2\bar{q}_2]$$

$$a_{22} = \left. \frac{\partial f_2}{\partial c} \right|_{eq.p.} = -\frac{\bar{q}_1}{S\bar{h}} - \frac{\bar{q}_2}{S\bar{h}}$$

Mixing tank (cont'd)

$$B = \left. \frac{\partial f(x,u)}{\partial u} \right|_{\bar{x}, \bar{u}} = \left[\begin{array}{cc} \frac{\partial f_1(x,u)}{\partial u_1} & \frac{\partial f_1(x,u)}{\partial u_2} \\ \frac{\partial f_2(x,u)}{\partial u_1} & \frac{\partial f_2(x,u)}{\partial u_2} \end{array} \right]_{\bar{x}, \bar{u}}$$

$$\dot{h} = (q_1 + q_2 - k\sqrt{h})/S \rightarrow f_1$$

$$\dot{c} = \frac{q_1}{Sh} (c_1 - c) + \frac{q_2}{Sh} (c_2 - c) \rightarrow f_2$$

$$b_{11} = \left. \frac{\partial f_1}{\partial q_1} \right|_{eq.p.} = \frac{1}{S}$$

$$b_{12} = \left. \frac{\partial f_1}{\partial q_2} \right|_{eq.p.} = \frac{1}{S}$$

$$b_{21} = \left. \frac{\partial f_2}{\partial q_1} \right|_{eq.p.} = \frac{c_1 - \bar{c}}{Sh}$$

$$b_{22} = \left. \frac{\partial f_2}{\partial q_2} \right|_{eq.p.} = \frac{c_2 - \bar{c}}{Sh}$$

Mixing tank (cont'd)

$$B = \left. \frac{\partial f(x,u)}{\partial u} \right|_{\bar{x}, \bar{u}} = \left[\begin{array}{cc} \frac{\partial f_1(x,u)}{\partial u_1} & \frac{\partial f_1(x,u)}{\partial u_2} \\ \frac{\partial f_2(x,u)}{\partial u_1} & \frac{\partial f_2(x,u)}{\partial u_2} \end{array} \right] \bigg|_{\bar{x}, \bar{u}}$$

$$\dot{h} = (q_1 + q_2 - k\sqrt{h})/S \rightarrow f_1$$

$$\dot{c} = \frac{q_1}{Sh} (c_1 - c) + \frac{q_2}{Sh} (c_2 - c) \rightarrow f_2$$

$$b_{11} = \left. \frac{\partial f_1}{\partial q_1} \right|_{eq.p.} = \frac{1}{S}$$

$$b_{12} = \left. \frac{\partial f_1}{\partial q_2} \right|_{eq.p.} = \frac{1}{S}$$

$$b_{21} = \left. \frac{\partial f_2}{\partial q_1} \right|_{eq.p.} = \frac{c_1 - \bar{c}}{S\bar{h}}$$

$$b_{22} = \left. \frac{\partial f_2}{\partial q_2} \right|_{eq.p.} = \frac{c_2 - \bar{c}}{S\bar{h}}$$

$$c_{11} = 1$$

$$c_{12} = 0$$

$$c_{21} = 0$$

$$c_{22} = \frac{1}{c_1}$$

Mixing tank (cont'd)

Residence time

At the operating point: $\dot{h} = (q_1 + q_2 - k\sqrt{h})/S = 0$

$$\theta := \frac{S\bar{h}}{\bar{q}_1 + \bar{q}_2} = \frac{S\sqrt{\bar{h}}}{k} \quad [\text{s}]$$

Linear(ized) model

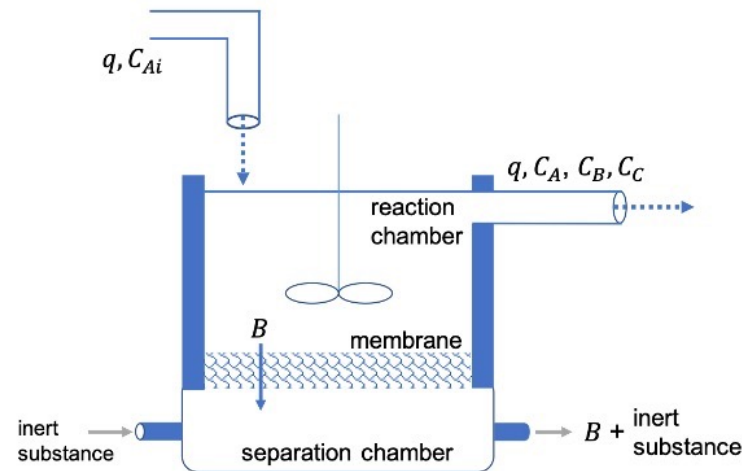
$$\begin{bmatrix} \delta \dot{h} \\ \delta \dot{c} \end{bmatrix} = \begin{bmatrix} -\frac{1}{2\theta} & 0 \\ 0 & -\frac{1}{\theta} \end{bmatrix} \begin{bmatrix} \delta h \\ \delta c \end{bmatrix} + \begin{bmatrix} \frac{1}{S} & \frac{1}{S} \\ \frac{c_1 - \bar{c}}{S\bar{h}} & \frac{c_2 - \bar{c}}{S\bar{h}} \end{bmatrix} \begin{bmatrix} \delta q_1 \\ \delta q_2 \end{bmatrix}$$

$$\begin{bmatrix} \delta h(0) \\ \delta c(0) \end{bmatrix} = \begin{bmatrix} h_0 - \bar{h} \\ c_0 - \bar{c} \end{bmatrix}$$

$$\begin{bmatrix} \delta y_1 \\ \delta y_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & \frac{1}{c_1} \end{bmatrix} \begin{bmatrix} \delta h \\ \delta c \end{bmatrix}$$

Exam problem example

1. (30 points) A reversible reaction $2A \leftrightarrow B + C$ takes place in a membrane reactor shown below. The reactor consists of a well-mixed reaction chamber and a separation chamber separated by a membrane. The membrane is permeable to compound B , whereas compounds A and C cannot pass through. Compound B diffuses out of the reaction chamber through the membrane and is removed from the separation chamber by a stream of inert substance. The diffusion rate of B through the membrane is $k_m A_m C_B [\frac{\text{kmol}}{\text{min}}]$, where A_m is the effective area of the membrane. Compound A is pumped into the reactor with the flow rate q and the concentration C_{Ai} . C_A , C_B , and C_C denote the concentrations of A , B , and C in the reaction chamber. Despite the diffusion of B through the membrane, the exit and feed flow rates are assumed equal because B is a small molecule, i.e., it is assumed that the reactor volume is constant. The temperature in the reactor is also constant. The reaction follows the mass action kinetics with the rate constants $k_1 = 0.5 \frac{\text{m}^3}{\text{kmol} \cdot \text{min}}$ (in the forward direction) and $k_2 = 2.5 \frac{\text{m}^3}{\text{kmol} \cdot \text{min}}$ (in the reverse direction).



For the parameter values: reactor volume $V = 3 \text{ m}^3$, feed flow rate $q = 0.25 \text{ m}^3/\text{min}$, mass transfer coefficient $k_m = 1.5 \text{ m/min}$, and input feed concentration $C_{Ai} = 10 \text{ kmol/m}^3$:

- Write the mass balances for compounds A and B .
- Knowing that the effective membrane area at the steady-state is $A_m^{ss} = 4 \text{ m}^2$, and that the steady-state concentration of compound A is five times the one of B , i.e., $C_A^{ss} = 5C_B^{ss}$, determine the steady-state concentrations of A , B , and C in the reaction chamber.
- The membrane degrades gradually and its effective area, A_m , changes in time. Derive the dynamic equations for deviations of concentrations C_A and C_B by linearizing the system of equations from a) around the steady-state derived in b).
- Having the matrices in the states-space form, discuss how would you express in the Laplace domain the dependency of C_A and C_B on A_m . No explicit calculation is needed.